

(1) Let variable  $\mathcal{X}$  follow a Bernoulli distribution with parameter  $p$ , i.e., the probability of  $x$  is  $f(x) = p^x(1-p)^{1-x}$ . If the set of independent and identically distributed data (i.i.d.) samples  $\mathcal{X} = \{x_1, \dots, x_N\}$  follows the Bernoulli distribution, derive the maximum likelihood estimate (MLE) for  $p^{MLE}$ .

$$\begin{aligned}
 \ell(p) &= \prod_{n=1}^N f(x_n) \\
 \mathcal{L}(p) &= \log \ell(p) = \log \prod_{n=1}^N f(x_n) = \sum_{n=1}^N \log f(x_n) = \\
 &= \sum_{n=1}^N \log [p^{x_n} (1-p)^{1-x_n}] = \sum_{n=1}^N [\log p^{x_n} + \log (1-p)^{1-x_n}] \\
 &= \sum_{n=1}^N [x_n \log p + (1-x_n) \log (1-p)] \\
 &= (\log p) \left( \sum_{n=1}^N x_n \right) + [\log (1-p)] \sum_{n=1}^N (1-x_n) \\
 \frac{\partial \mathcal{L}(p)}{\partial p} &= \frac{1}{p} \left( \sum_{n=1}^N x_n \right) + \left( -\frac{1}{1-p} \right) \sum_{n=1}^N (1-x_n) = \frac{(1-p) \sum_{n=1}^N x_n - p \sum_{n=1}^N (1-x_n)}{p(1-p)} = \\
 &= \frac{\sum_{n=1}^N x_n - p \sum_{n=1}^N x_n - pN + p \sum_{n=1}^N x_n}{p(1-p)} = \frac{\sum_{n=1}^N x_n - pN}{p(1-p)} \\
 \frac{\partial \mathcal{L}(p)}{\partial p} = 0 &\Rightarrow \frac{\sum_{n=1}^N x_n - pN}{p(1-p)} = 0 \Rightarrow \boxed{\hat{p} = \frac{1}{N} \sum_{n=1}^N x_n}
 \end{aligned}$$

(2) Let variable  $\mathcal{X}$  follow an exponential distribution with parameter  $\lambda$ , i.e., the probability of  $x$  is  $f(x) = \lambda e^{-\lambda x}$ . If the set of independent and identically distributed data (i.i.d.) samples  $\mathcal{X} = \{x_1, \dots, x_N\}$  follows the exponential distribution, derive the maximum likelihood estimate (MLE) for  $\lambda^{MLE}$ .

$$\begin{aligned}
 \ell(\lambda) &= \prod_{n=1}^N f(x_n) \\
 \mathcal{L}(\lambda) &= \log \prod_{n=1}^N f(x_n) = \sum_{n=1}^N \log f(x_n) = \sum_{n=1}^N \log (\lambda e^{-\lambda x_n}) \\
 &= \sum_{n=1}^N (\log \lambda + \log e^{-\lambda x_n}) = \sum_{n=1}^N (\log \lambda - \lambda x_n) \\
 &= N \cdot \log \lambda - \lambda \sum_{n=1}^N x_n \\
 \frac{\partial \mathcal{L}(\lambda)}{\partial \lambda} &= \frac{N}{\lambda} - \sum_{n=1}^N x_n = 0 \Rightarrow \boxed{\hat{\lambda} = \frac{1}{N} \sum_{n=1}^N x_n}
 \end{aligned}$$